

Biologically Inspired Computing

EECS 6180

Homework 3

Hill Climbing

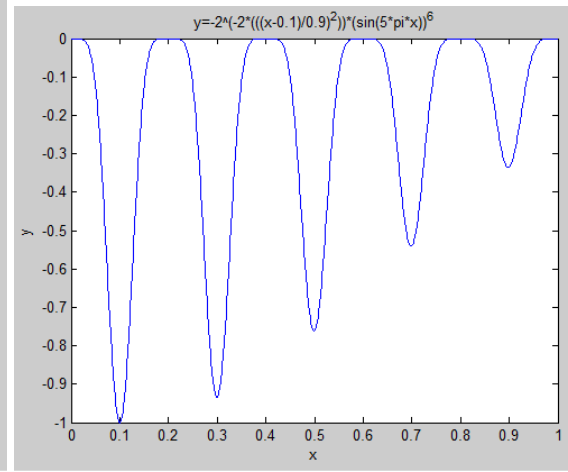
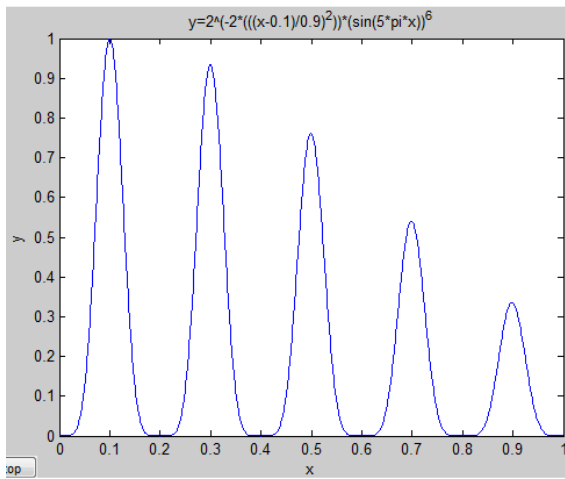
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- Plot of the Function

$$y = 2^{-2\left(\frac{x-0.1}{0.9}\right)^2} * (\sin(5\pi x))^6$$

$$y = -2^{-2\left(\frac{x-0.1}{0.9}\right)^2} * (\sin(5\pi x))^6$$



For using gatool the function needed to be inverted because gatool's fitness function locates optimum values which give minimum output. The plot for the function in the fitness function is shown in the figure above on the right side.

- The fitness function you write for GA tool.

```
function y = fitness1(x)
    y=-2^(-2*((x-0.1)/0.9)^2)*(sin(5*pi*x))^6;
end
```

For this assignment I tested different variations to understand the effect of Selection, Crossover Probability, and Mutation Probability. For each test I ran ten tests and generated tables showing the relationship between convergence and one of the parameters listed earlier. In all cases the error was calculated as the distance from the final point and 0.1 the true global optimum value.

$$Error = (Final Point) - 0.1$$

Each variation was run 10 iterations to obtain the average error to compare the overall effects of the different probabilities and selection techniques.

$$Average Error = \frac{\sum E}{10}$$

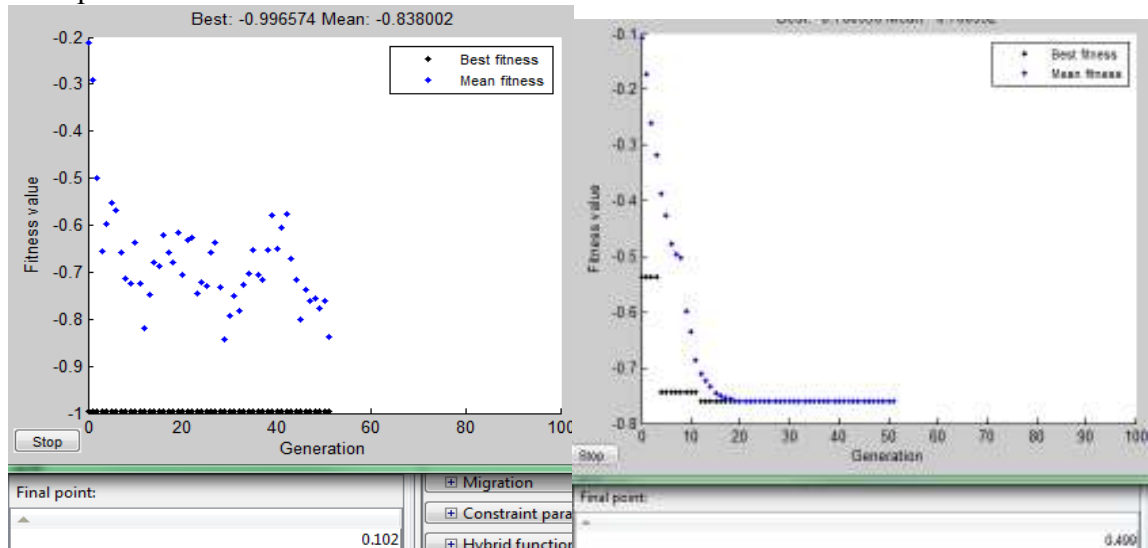
- The screen capture of the final result which returns the optimum value for x.
Convergence graph

Without any changes made to the default gatool settings:

Test 1: Selection: Stochastic Uniform

Pop. Size:20 Pcrossover: 0.8 Pmutation: N/A Stalling: 50 Generations

Final point: 0.102 0.499



Iter. #	Final Point	Error = (Final Point) - 0.1	Average Error
1	0.102	0.001	0.999/10 = 0.0999
2	0.499	0.399	
3	0.1	0	
4	0.1	0	
5	0.1	0	
6	0.299	0.199	
7	0.1	0	
8	0.3	0.2	
9	0.1	0	
10	0.3	0.2	

Reducing the population size by half caused gatools to converge on local optima and sometimes converged to the global optima.

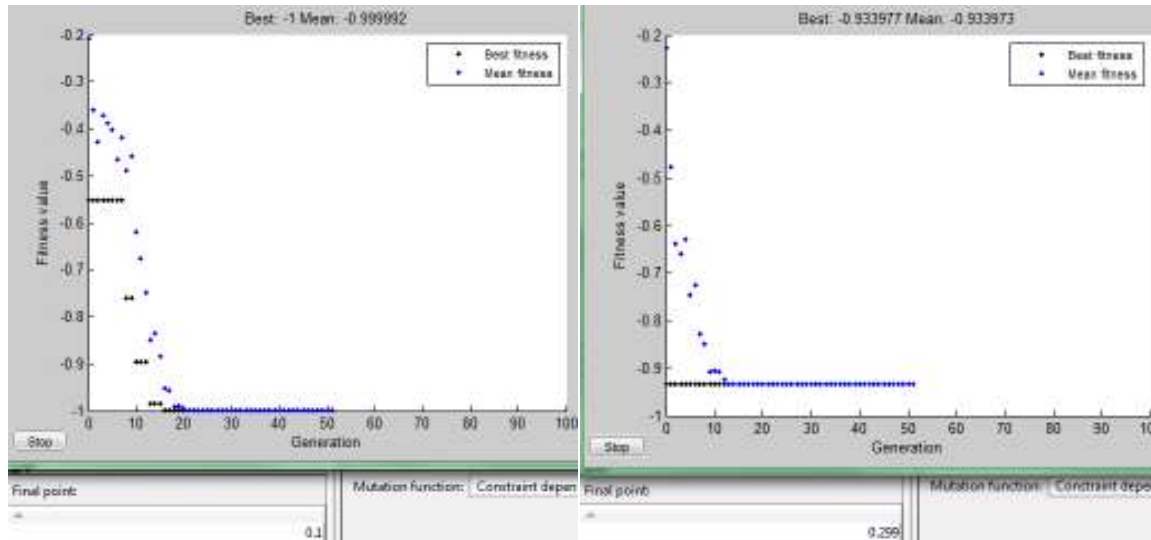
Test 2: Selection: Stochastic Uniform

Pop. Size: 10 Pcrossover: 0.8 Pmutation: N/A

Stalling: 50 Generations

Final Point: 0.1

0.299



Iter. #	Final Point	Error = (Final Point) - 0.1	Average Error
1	0.1	0	2.194/10 =0.2914
2	0.299	0.199	
3	0.499	0.399	
4	0.699	0.599	
5	0.3	0.2	
6	0.1	0	
7	0.1	0	
8	0.299	0.199	
9	0.299	0.199	
10	0.499	0.399	

Based on information from the table above, reducing the population size leads to a larger error in convergence.

Now let us see the effect of changing the selection function to Roulette and keeping all other settings as the previous test.

Test 3: Selection: Roulette

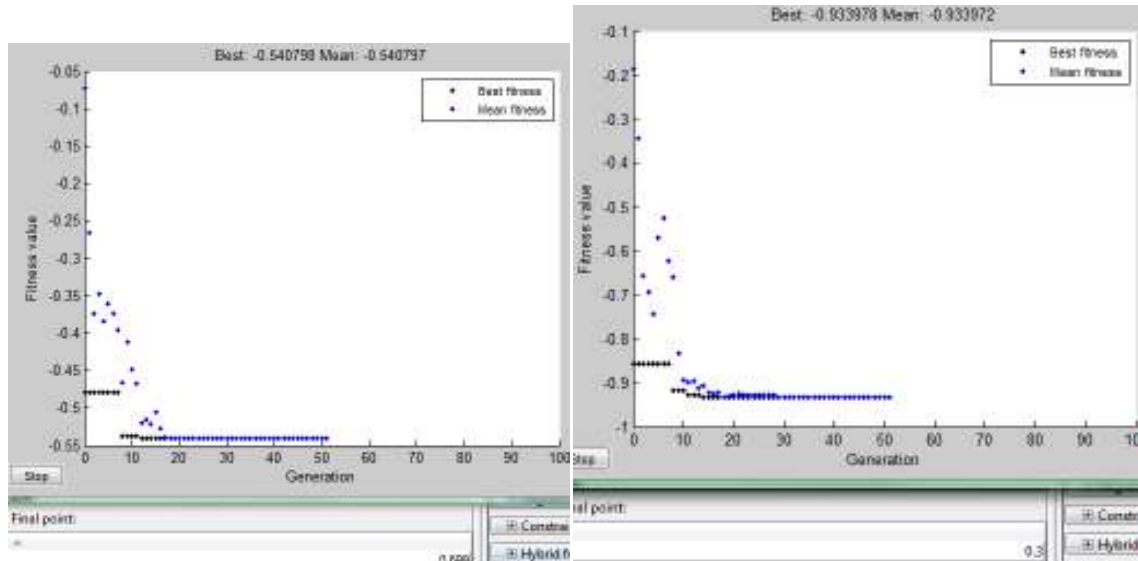
Pop. Size: 10 Pcrossover: 0.8

Pmutation: N/A

Stalling: 50 Generations

Final Point: 0.699

0.3



Iter. #	Final Point	Error = (Final Point) - 0.1	Average Error
1	0.699	0.599	3.193/10 =0.3913
2	0.3	0.2	
3	0.1	0	
4	0.699	0.599	
5	0.1	0	
6	0.299	0.199	
7	0.499	0.399	
8	0.1	0	
9	0.898	0.798	
10	0.499	0.399	

For this test I will be observing the effect of reducing the probability of crossover to 0.5.

Test 4. Selection: Roulette

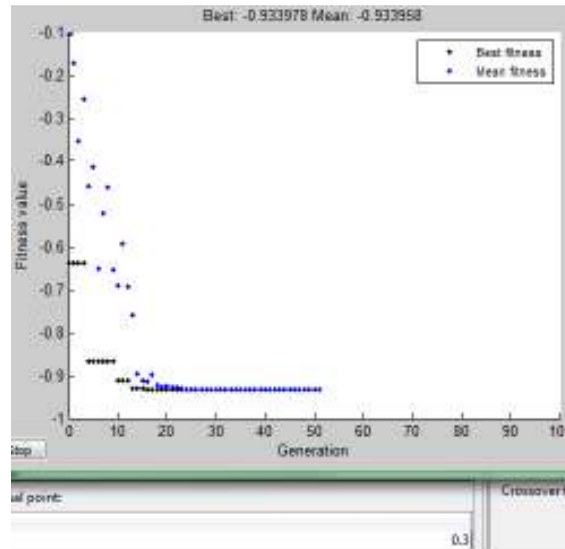
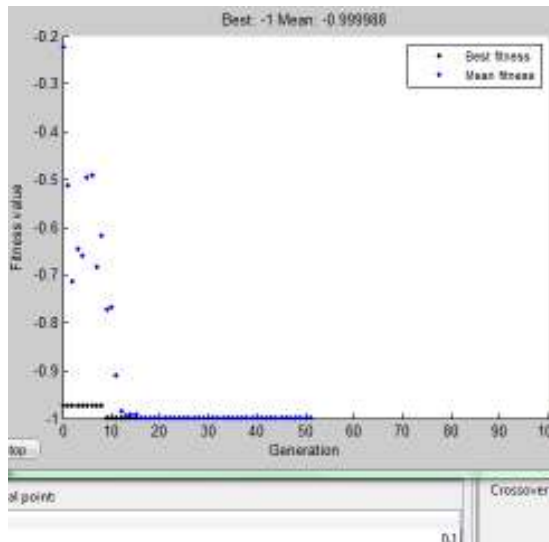
Pop. Size: 10 **Pcrossover: 0.5**

Pmutation: N/A

Stalling: 50 Generations

Final Point: 0.1

0.3



Iter. #	Final Point	Error = (Final Point) - 0.1	Average Error
1	0.1	0	1.798/10 =0.1798
2	0.3	0.2	
3	0.699	0.599	
4	0.499	0.399	
5	0.1	0	
6	0.1	0	
7	0.1	0	
8	0.3	0.2	
9	0.3	0.2	
10	0.3	0.2	

The next table was generated to compare the results of Roulette selection and Stochastic selection with $P_c = 0.5$

Test 5. Selection: **Stochastic**

Pop. Size: 10 **Pcrossover: 0.5**

Pmutation: N/A

Stalling: 50 Generations

Iter. #	Final Point	Error = (Final Point) - 0.1	Average Error
1	0.499	0.399	1.597/10 =0.1597
2	0.1	0	
3	0.1	0	
4	0.299	0.199	
5	0.3	0.2	
6	0.3	0.2	
7	0.3	0.2	
8	0.1	0	
9	0.1	0	
10	0.499	0.399	

Now I will observe the effect of mutation by first setting the Probability of Uniform mutation to 0.01.

Test 6. Selection: Roulette

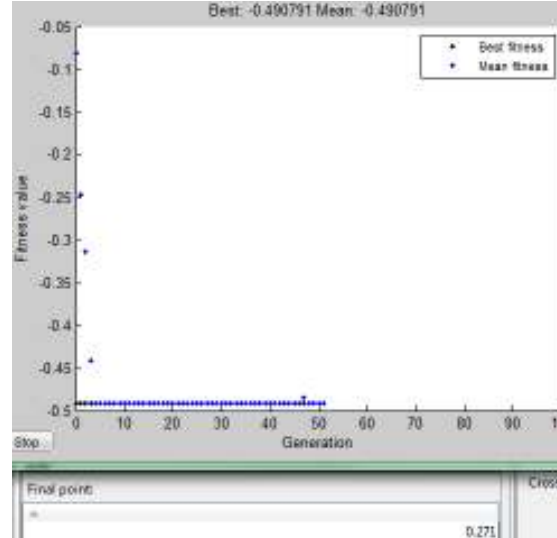
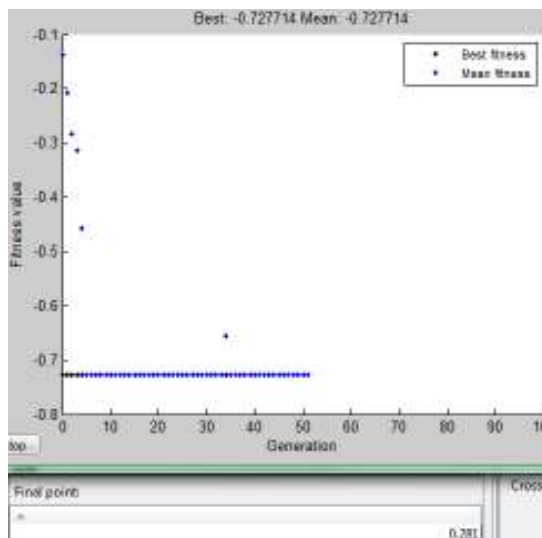
Pop. Size: 10 Pcrossover: 0.8

Pmutation: 0.01

Stalling: 50 Generations

Final Point: 0.281

0.271



Iter. #	Final Point	Error = (Final Point) - 0.1	Average Error
1	0.281	0.181	1.501/10 =0.1501
2	0.271	0.171	
3	0.3	0.2	
4	0.298	0.198	
5	0.091	-0.009	
6	0.262	0.162	
7	0.292	0.192	
8	0.12	0.02	
9	0.3	0.2	
10	0.286	0.186	

The following data is the results from increasing the Probability of mutation to 0.1 and keeping all other setting the same as the previous test.

Test 7. Selection: Roulette

Pop. Size: 10 Pcrossover: 0.8

Pmutation: 0.1

Stalling: 50 Generations

Iter. #	Final Point	Error = (Final Point) - 0.1	Average Error
1	0.099	-0.001	1.594/10 =0.1594
2	0.312	0.212	
3	0.699	0.599	
4	0.097	-0.003	
5	0.087	-0.013	
6	0.299	0.199	
7	0.308	0.208	
8	0.28	0.18	
9	0.299	0.199	
10	0.114	0.014	

In the following test all parameters were kept the same but the Probability of mutation was significantly increased to 0.5

Test 8. Selection: Roulette

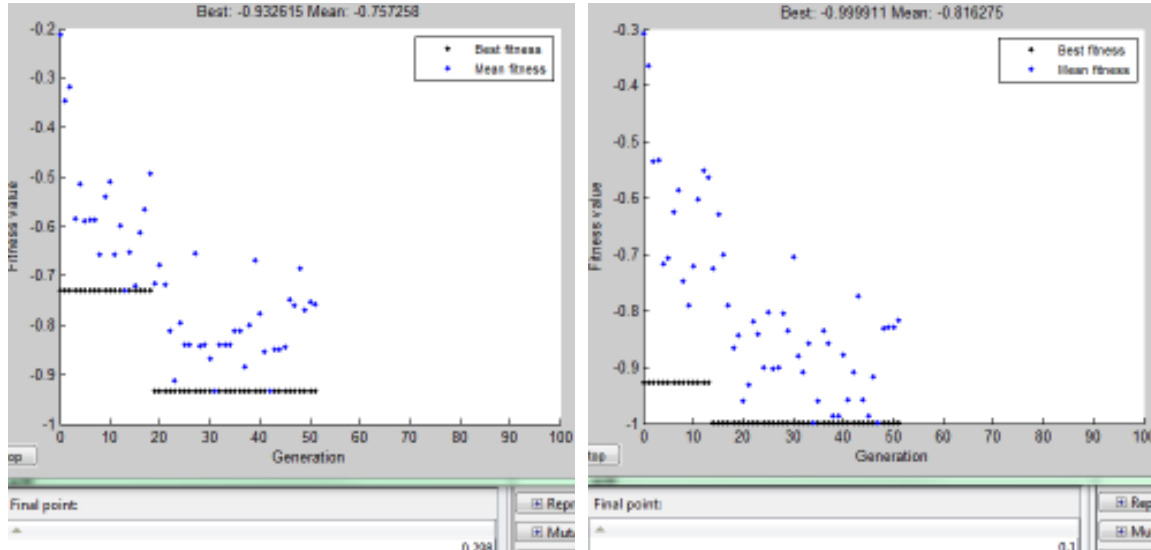
Pop. Size: 10 Pcrossover: 0.8

Pmutation: 0.5

Stalling: 50 Generations

Final Point: 0.298

0.1



Iter. #	Final Point	Error = (Final Point) - 0.1	Average Error
1	0.298	0.198	0.976/10 =0.0976
2	0.1	0	
3	0.1	0	
4	0.101	0.001	
5	0.292	0.192	
6	0.1	0	
7	0.302	0.202	
8	0.1	0	
9	0.296	0.196	
10	0.287	0.187	

With the same settings as above the table shows the effect of probability of crossover

Test 9. Selection: Roulette

Pop. Size: 10 Pcrossover: 0.5

Pmutation: 0.5

Stalling: 50 Generations

Iter. #	Final Point	Error = (Final Point) - 0.1	Average Error
1	0.095	-0.005	0.396/10 =0.0396
2	0.101	0.001	
3	0.105	0.005	
4	0.101	0.001	
5	0.092	-0.008	
6	0.298	0.198	
7	0.105	0.005	
8	0.091	-0.009	
9	0.298	0.198	

Conclusion

The table below summarizes the variations in all 9 tests and the average error.

-From the table it can be seen that a larger population size converges with less error.

Test 1 VS Test 2

-The table shows that less crossover probability leads to less error

Test 8 VS Test 9

Test 3 VS Test 4

Test 2 VS Test 5

-The table shows that significantly greater uniform mutation probability along with Roulette selection leads to less error

Test 7 VS Test 8

- Table summarizing the performance of you choices

Test #	Population Size	Selection Function	Probability Crossover	Probability Mutation	Average Error
1	20	STOCHASTIC	0.8	NA	0.0999
2	10	STOCHASTIC	0.8	NA	0.2919
3	10	ROULETTE	0.8	NA	0.3913
4	10	ROULETTE	0.5	NA	0.1798
5	10	STOCHASTIC	0.5	NA	0.1597
6	10	ROULETTE	0.8	0.01	0.1501
7	10	ROULETTE	0.8	0.1	0.1594
8	10	ROULETTE	0.8	0.5	0.0976
9	10	ROULETTE	0.5	0.5	0.0396